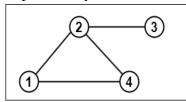
Mid-term Exam (Graph Neural Networks –Fall 2023)

Full Name:

Student ID:

- 1. Consider an undirected graph G of four nodes given in the following figure. The normalized adjacency matrix is defined as: $\tilde{A} = D^{-1} \cdot A$, where A is the adjacency matrix of G and D is the node degree matrix.
 - a) Calculate the matrix A
 - b) Calculate the 2-step transition probability matrix \tilde{A}^2



Solutions:

$$A = \begin{bmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \end{bmatrix} \qquad D = \begin{bmatrix} 2 \\ 3 \\ 1 \\ 2 \end{bmatrix}$$

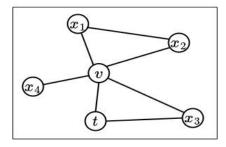
$$\tilde{A} = \frac{A}{D} = \begin{bmatrix} 0 & .5 & 0 & .5 \\ .33 & 0 & .33 & .33 \\ 0 & 1 & 0 & 0 \\ .5 & .5 & 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & .5 & 0 & .5 \end{bmatrix}$$

$$\tilde{A} = \frac{A}{D} = \begin{bmatrix} 0 & .5 & 0 & .5 \\ .33 & 0 & .33 & .33 \\ 0 & 1 & 0 & 0 \\ .5 & .5 & 0 & 0 \end{bmatrix}$$

$$\tilde{A}^2 = \tilde{A} \times \tilde{A} = \begin{bmatrix} 0 & .5 & 0 & .5 \\ .33 & 0 & .33 & .33 \\ 0 & 1 & 0 & 0 \\ .5 & .5 & 0 & 0 \end{bmatrix} \times \begin{bmatrix} 0 & .5 & 0 & .5 \\ .33 & 0 & .33 & .33 \\ 0 & 1 & 0 & 0 \\ .5 & .5 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0.42 & 0.25 & 0.17 & 0.17 \\ 0.17 & 0.67 & 0 & 0.17 \\ 0.33 & 0 & 0.33 & 0.33 \\ 0.17 & 0.25 & 0.17 & 0.42 \end{bmatrix}$$

2. Consider the Node2vec algorithm with the return parameter p = 0.2 and the in-out parameter q = 0.6. A walker traversed from node t to node v and now resides at node v. Calculate the transition probabilities from node v to other nodes.



Solutions:

$$P(v, x_1) = \frac{1}{q} = \frac{1}{0.6}$$

$$P(v, x_2) = \frac{1}{q} = \frac{1}{0.6}$$

$$P(v, x_3) = 1$$

$$P(v, x_4) = \frac{1}{q} = \frac{1}{0.6}$$

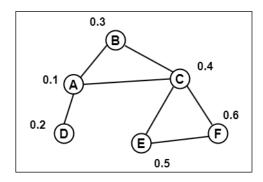
$$P(v,t) = \frac{1}{p} = \frac{1}{0.2}$$

3. Consider an undirected graph G of six nodes A, B, C, D, E, and F given in the following figure. Each node has initial features that are the numbers standing next to it (i.e., the initial feature of node 'A' is $X_A^{(0)} = 0.1$). Assume that the feature of a node *i* at layer *k* can be updated as:

$$X_i^{(k)} = F(X_i^{(k-1)}, N_i^{(k-1)}),$$

where $F(\bullet)$ returns a numeric value that is the mean (average) of its arguments, $N_i^{(k-1)}$ is the set of features of the neighbours of node i at layer (k-1).

- a) Calculate the feature of each node at k = 1.
- b) Calculate the feature of each node at k = 2.



Solutions:

$$X_{A}^{1} = mean\left(X_{A}^{0} + X_{B}^{0} + X_{C}^{0} + X_{D}^{0}\right) = \frac{0.1 + 0.2 + 0.3 + 0.4}{4} = 0.25$$

$$X_{B}^{1} = mean\left(X_{A}^{0} + X_{B}^{0} + X_{C}^{0}\right) = (0.1 + 0.3 + 0.4)/3 = 0.27$$

$$X_{C}^{1} = (0.1 + 0.3 + 0.4 + 0.5 + 0.6)/5 = 0.38$$

$$X_{D}^{1} = (0.1 + 0.2)/2 = 0.15$$

$$X_{E}^{1} = (0.4 + 0.5 + 0.6)/3 = 0.5$$

$$X_{E}^{1} = (0.4 + 0.5 + 0.6)/3 = 0.5$$

$$X_{A}^{2} = mean\left(X_{A}^{1} + X_{B}^{1} + X_{C}^{1} + X_{D}^{1}\right) = \frac{0.25 + 0.27 + 0.38 + 0.15}{4} = 0.26$$

$$X_{B}^{2} = mean\left(X_{A}^{1} + X_{B}^{1} + X_{C}^{1}\right) = (0.25 + 0.27 + 0.38) / 3 = 0.3$$

$$X_{C}^{2} = (0.25 + 0.27 + 0.38 + 0.5 + 0.5) / 5 = 0.38$$

$$X_{D}^{2} = (0.25 + 0.15) / 2 = 0.2$$

$$X_{E}^{2} = (0.38 + 0.5 + 0.5) / 3 = 0.43$$

$$X_{E}^{2} = (0.38 + 0.5 + 0.5) / 3 = 0.46$$

4. Given a graph with an adjacency matrix A and initial node feature matrix $H^{(0)}$ as follows:

$$A = \begin{bmatrix} 0 & 1 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 0 \end{bmatrix} \qquad H^{(0)} = \begin{bmatrix} 0 & 0 \\ 1 & -1 \\ 2 & -2 \\ 3 & -3 \\ 4 & -4 \\ 5 & -5 \end{bmatrix}$$

Assume that the hidden layer of an GCN model of all nodes at layer (k) can be calculated as:

$$H^{(k)} = \sigma(A \cdot H^{(k-1)}),$$

where $H^{(k)}$ denotes the output at layer k, σ is a ReLU function ReLU(x) = max(0, x). Calculate the output of the GCN model at layer k = 1.

Solutions:

$$H^{1} = \sigma(AH^{0}) = \text{ReLU} \begin{bmatrix} 0 & 1 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 0 \end{bmatrix} \times \begin{bmatrix} 0 & 0 \\ 1 & -1 \\ 2 & -2 \\ 3 & -3 \\ 4 & -4 \\ 5 & -5 \end{bmatrix} = \text{ReLU} \begin{bmatrix} 6 & -6 \\ 2 & -2 \\ 1 & -1 \\ 9 & -9 \\ 8 & -8 \\ 7 & -7 \end{bmatrix} = \begin{bmatrix} 6 & 0 \\ 2 & 0 \\ 1 & 0 \\ 9 & 0 \\ 8 & 0 \\ 7 & 0 \end{bmatrix}$$